Path-search in the pyramid and in other graphs

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Abstract

We are given an acyclic directed graph with one source, and a subset of its edges which contains exactly one outgoing edge for every non-sink vertex. These edges determine a unique path from the source to a sink. We can think of it as a switch in every vertex, which determines which way the water arriving to that vertex flows further.

We are interested in determining either the sink the flow arrives, or the whole path, with as few questions as possible. The questions we can ask correspond to the vertices of the graph, and the answer describes the switch, i.e. tells which outgoing edge is in our given subset.

Originally the problem was proposed by Soren Riis (who posed the question for pyramid graphs) in the following more general form. We are given a natural number k, and k questions can be asked in a round. The goal is to minimize the number of rounds. We completely solve this problem for complete t-ary trees. Also, for pyramid graphs we present some non-trivial partial results.

1 Introduction

In this paper we consider the following problem. We are given a directed acyclic (multi)graph G with one source. We are given a set $E^* \subset E$ such that for every non-sink vertex $v \in V$ there is exactly one outgoing edge in E^* . It determines a path from the source to a sink. We can think of it as a switch in every vertex, which determines which way the water (coming from the source) arriving to that vertex flows further. We imagine such graphs such that the source is on the top and the sinks are on the bottom. In the rest of the paper by a graph we always mean such a directed graph except when noted.

We are interested in determining either the whole path $P = P(E^*)$ or only the sink the flow arrives, i.e. the last vertex of $P(E^*)$. The questions we can ask correspond to the vertices of the graph, and the answer describes the switch, i.e. tells which outgoing edge is in E^* .

Definition 1.1. Let $pa(G) = pa_1(G)$ (resp. $si(G) = si_1(G)$) be the minimal number of questions we need to determine the path (resp. the sink).

Clearly $si(G) \leq pa(G)$, but these can be very far from each other, for example if there is only one sink, then si(G) = 1 but pa(G) can be arbitrarily high.

Soren Riis proposed the following problem. What is the minimal number of rounds we need if in one round we can ask k questions and our aim is to determine the path or the sink in the pyramid graph (for the definition see Section 4)? This motivates the study of the following more general question.

Definition 1.2. Fixing a k, $pa_k(G)$ is the minimal number of rounds we need to determine the path, if in each round we can ask k questions. Similarly, $si_k(G)$ is the minimal number of rounds we need to determine the sink, if in each round we can ask k questions.

As an additional motivation, let us consider a game and two possible strategies given as black boxes, i.e. we can evaluate in every possible state what the strategies do. To represent the game with a directed acyclic graph, it is enough represent every valid (turn number, state) pair with a different vertex, and the valid steps are represented by the directed edges. For every finite game, this is a finite acyclic directed graph. Now we can simulate a match between two players using the respective strategies. To find out which strategy is the winner of a match, one can go step by step and ask what happens in every actual situation. Note that it does not matter if the black box strategies use randomness or not. This process is a very important step of Monte-Carlo type algorithms, when one needs to quickly simulate matches between strategies [1, 2, 3].

However, suppose we are given multiple processors. A solution to the original sink-search problem for the above defined graph gives an optimal parallel algorithm to determine the final state of the match. Clearly, such an algorithm determines the winner of the match as well. Summarizing, investigating our problem can help find faster algorithms for simulating matches, if parallel computations are allowed. (We note that we do not think it could help a lot. We were able to find an example where the optimal algorithm is much faster than the trivial one, but it was a very special graph. Also in case of games usually the graph is very large, which makes it hard to find a good non-trivial algorithm for our problem.) For further motivations related to random walks see the beginning of Section 4.

In the next section we examine si(G) and pa(G) in directed acyclic graphs. In Section 3 we consider $si_k(G)$ and $pa_k(G)$ on trees and completely solve the problem on d-ary trees. In Section 4 we consider the problem on the pyramid graph. In the last section we conclude our paper with some additional remarks and open problems.

2 Search in directed acyclic graphs

We will show a process how to transform the graph G into a graph G' such that si(G) = si(G') and si(G') can be easily determined. We also show how to transform G into a graph G'' such that pa(G) = pa(G'') and pa(G'') can be easily determined.

But at first let us consider a trivial algorithm which finds both the sink and the path: at first the source is asked. Then the answer tells us which vertex is the next on the path. That one is asked in the next turn, and so on. This gives us the following simple observation.

Observation 2.1. For every (multi)graph G, $si(G) \le pa(G) \le l(G)$, where l(G) is the length (the number of edges) of the longest path.

We start with examining si(G). We can simply forget about multiple edges then, hence we suppose G is simple. Obviously it's useless to ask a vertex with outdegree 1. We will define a graph G' with no such vertices such that if we find the sink in G' it gives us the sink in G (in fact it is the same vertex). We introduce the following merging operation: for a set of vertices M, we get G(M) from G by deleting the vertices M and introducing a new vertex m, if there was an edge between a vertex v in $G \setminus M$ and a vertex w in M then we put an edge in the new graph between v and m with the same orientation as in G. If multiple edges come into existence, we consider them as one edge. Edges between vertices of $G \setminus M$ stay untouched. Acyclicity could be ruined by such an operation but anytime we do such an operation, it will be easy to see that acyclicity remains true.

If a vertex x of G has exactly one out-neighbor y, then $G(\{x,y\})$ remains acyclic and $si(G(\{x,y\})) = si(G)$. By merging this way vertices with outdegree 1 with their outneighbors as long as it is possible, we get a graph G' with vertices all having outdegree minimum 2 (except the sinks) and for which si(G') = si(G). Although it is not relevant for us, it is easy to see that G' does not depend on the order in which we process the vertices. The merging procedure defines a map from the vertices of G to the vertices of G', hopefully without causing confusion, we will refer to the image of some vertex x of G also as x. By the procedure, every vertex of G' has at least 2 out-neighbors or it is a sink.

Observation 2.2. If a vertex y can be reached from a vertex x in G, then y can be reached from x in G' as well.

The main result of this section is that after merging every vertex with outdegree 1 to get G', si(G') can be determined easily:

Theorem 1. Suppose there is no vertex with outdegree 1 in a simple graph G. Then si(G) = l(G), where l(G) is the length (the number of edges) of the longest path.

Proof. We need some preliminary observations. Let us examine what happens when a question is answered. Clearly $xy \in E^*$ is equivalent to the following: all the other edges starting from x are not in E^* . But then we can simply delete these edges to get a new graph and then the outdegree of x becomes 1, hence we can merge x with y to get another graph which we denote by G_{xy} , ie. this answer reduced the problem to finding the sink in G_{xy} , which is the graph we get by deleting from G all the edges going out from x except xy and then merging x and y. Thus, if the first question is x and the answer is xy, then $si(G_{xy})$ additional questions are needed to find the sink.

Definition 2.3. A directed path is called full if it ends in a sink.

Let us consider G'_{xy} . By definition x and y are merged. Then if a vertex z had only two outneighbors, x and y, it gets merged with them. Then if all the outneighbors of another vertex are among x, y and z it also gets merged with them, and so on. Let M be the set of vertices M of G for which every full path starting at some vertex u in M contains x or y. By the above argument, for this M, $G'_{xy} = G_{xy}(M)$.

Observation 2.4. If in G there is a full path starting at some vertex z and avoiding x and y, then when merging x and y, z cannot be among the merged vertices M.

Note that if G does not have vertices with outdegree 1, then if a vertex m is among the merged vertices M, then there are paths to both x and y from u, hence by acyclicity of G there are no paths from x to u.

Observation 2.5. If there is a path from x to z in G, then when merging x and y, z cannot be among the merged vertices M.

Now we can start to prove the theorem.

By Observation 2.1 $si(G) \le pa(G) \le l(G)$, hence it is enough to prove $l(G) \le si(G)$. We prove it by induction on k = l(G). The case k = 1 can be easily seen.

Now we describe the strategy of the adversary. Let P be a path of length k, consisting of the edges $x_1x_2, x_2x_3, \ldots, x_kx_{k+1}$.

Case 1. Suppose that the first vertex asked is x_i , and there is an edge $x_{i-1}x_{i+1}$ in the graph. Then the adversary's answer should be an edge x_iy , where $y \neq x_{i+1}$. It means we continue with the graph G_{x_iy} . It might contain vertices with outdegree 1, hence we need to determine G'_{x_iy} .

By Observation 2.4 the vertices x_j $(j \neq i)$ do not get merged in G'_{x_iy} , thus there is a path of length k-1 in G'_{x_iy} , containing $x_1, \ldots, x_{i-1}, x_{i+1}, \ldots, x_{k+1}$ in this order. Then by induction at least k-1 additional questions are needed in G'_{x_iy} .

Case 2. The first vertex asked is x_i but there is no edge $x_{i-1}x_{i+1}$ in the graph G. Then the adversary's answer should be the edge x_ix_{i+1} . No vertex x_j with $j \geq i$ gets merged by Observation 2.5. Also, from x_{i-1} there is an edge going out which is not the edge $x_{i-1}x_i$. Going

along this edge we can find a path Q. First we claim that for any choice of such a Q, it does not contain x_i . Indeed, otherwise the original path P minus the edge $x_{i-1}x_i$ plus this path Q would give a longer path in G, a contradiction. Further, Q can be chosen such that it avoids x_{i+1} as well. Indeed, every outdeegree is at least 2, so we always have at least one choice different from x_{i+1} to continue. This way, for every vertex x_j with j < i there is a full path starting at x_j and avoiding both x_i and x_{i+1} , thus by Observation 2.4 none of these vertices gets merged in $G'_{x_ix_{i+1}}$. Hence the path $x_1, \ldots, x_{i-1}, x_{i+1}, \ldots, x_{k+1}$ is a k-1 long path in $G'_{x_ix_{i+1}}$, the induction can be applied.

Case 3. The first vertex asked, x is not in the path P. If it has an outneighbor y not on the path, that should be the adversary's answer. Then clearly the full path P is in G'_{xy} and by Observation 2.4 none of its vertices are merged. Hence by induction k more questions are needed.

Thus we can suppose that all the outneighbors are on P. Let x_i be the first and x_j be the last among them. Then the answer of the adversary is x_j . Again, by Observation 2.5, no x_l with l > j gets merged. Similarly as in Case 2, any full path starting at x_i must avoid x and also we can choose such a path which avoids x_j as well. This path Q shows by Observation 2.4 that x_i won't be merged. By adding to Q the appropriate part of the path P, for any x_l , l < i we can build a path avoiding both x and x_j , thus showing that these vertices won't be merged neither. Finally, for a vertex x_l with i < l < j, x_l can be reached by a directed path from x (starting with the edge xx_i and then going along P), thus by Observation 2.5 such a vertex cannot be merged neither.

As none of the vertices of the path are merged in G'_{xx_j} , by induction k more questions are needed.

Corollary 2.6. For any (multi)graph G, si(G) = si(G') = l(G').

Theorem 1 implies also that for a simple graph G, pa(G) = l(G), if all non-sink vertices of G have outdegree at least 2. We will prove that this holds even for multi-graphs.

We examine the path-search problem for multigraphs, hence we can claim again: it's useless to ask a vertex with outdegree 1. We introduce the following modified merging operation for multigraphs: for a set of vertices M, we get G[M] from G by deleting the vertices M and introducing a new vertex m, if there was an edge between a vertex v in $G \setminus M$ and a vertex w in M then we put an edge in the new graph between v and m with the same orientation as in G. Now, differently from the previous merging operation, if parallel edges come into existence, we consider them as parallel edges. Edges between vertices of $G \setminus M$ stay untouched.

Merging a vertex x with outdegree 1 with its outneighbor y to get $G[\{x,y\}]$, the graph remains acyclic and $pa(G[\{x,y\}]) = pa(G)$. By merging this way vertices with outdegree one with their outneighbors as long as it is possible, we get a graph G'' with vertices all having outdegree minimum 2 (except the sinks) and for which pa(G'') = pa(G).

Theorem 2. Suppose there is no vertex with outdegree 1 in a (multi)graph G. Then pa(G) = l(G).

Proof. We get $pa(G) \leq l(G)$ from Observation 2.1 so we only have to deal with the other direction. We examine what happens when a question is answered. Having $xy \in E^*$ is still equivalent to the following: all the other edges starting from x are not in E^* . But then we can simply delete these edges to get a new graph and then the outdegree of x becomes 1, hence we can use our new merging operation $G[\{x,y\}]$ to get another graph which we denote by G^{xy} , ie. this answer reduced the problem to finding the sink in G^{xy} , which is the graph we get by deleting from G all the edges going out from x except xy and then merging x and y. Thus, if the first question is x and the answer is xy, then $pa(G^{xy})$ additional questions are needed to find the sink. Note that $(G^{xy})'' = G^{xy}$. Now, if the asked vertex is from the fixed maximal path P, then the adversary answers the edge of the path, otherwise it answers arbitrarily. Thus, after any question and answer only at most one edge of P gets merged, thus the length of the maximal path reduces by at most one. From this by induction it easily follows that the size of the maximal path is a lower bound to the number of questions needed.

Corollary 2.7. For any (multi)graph G, pa(G) = pa(G'') = l(G'').

Another interesting question is the case when the graph contains a directed cycle. In this case it can happen that following the flow we get stucked in a directed cycle and never go to a sink.

We begin with the path-search problem. Let us call a set of edges a generalized path if it contains a directed path and possibly an additional edge which goes from the last vertex to a vertex already on the path. Let l'(G) denote the length (number of edges) of the longest generalized path starting at the start vertex.

One can easily see that if a multigraph G does not contain any vertices with outdegree 1, then pa(G) = l'(G). Indeed, we can copy the proof of Theorem 2. Consider a graph G and a generalized path P of length l'(G). No matter what the question and the answer are, at most two vertices are merged, hence at least l'(G) - 1 vertices of P remain. Again, the adversary answers in a way that for the vertices of the fixed maximal generalized path P the answer is always an edge of P. Then the remaining vertices of P form a generalized path of length at least l'(G) - 1, and the induction can be applied. Similarly, Corollary 2.7 remains true as well.

On the other hand, in case of si(G) we encounter problems as our basic operation for handling an answer cannot be applied, as we might lose some information at every merging. For example let us suppose z has only two outneighbors, x and y. Then z gets merged with them in G_{xy} . However, if z, x and y are part of the cycle where the flow ends in this order, then after merging x and y there is no way to differentiate this cycle from the other cycle where y comes immediately after z.

3 Search in trees

We consider trees as rooted directed trees where the edges are directed away from the root. For complete d-ary trees on n + 1 levels (denoted by $T_d(n)$) the obvious algorithm that asks as many complete levels as possible from the beginning, is the best possible:

Theorem 3. $pa_k(T_d(n)) = si_k(T_d(n)) = \lceil n/\log_d' k \rceil$, where \log_d' is defined as the biggest integer i for which $1 + d + d^2 + \ldots + d^{i-1} \le k$ holds.

Proof. Clearly, if we ask in each round the first as many full levels as possible, in each round P_{max} gets longer by \log'_d in worst case, thus in $\lceil n/\log'_d k \rceil$ rounds we can easily find the path. We give a strategy to the adversary, so that his answers force that from k queries of any round, only at most \log'_d will be on the final path P. This way, as the final path has length n, there were indeed at least $\lceil n/\log'_d k \rceil$ queries altogether.

Now fix a round i, define S as the set of vertices asked in this round. Now take a vertex $v \in S$ asked in this round. This vertex has d children, taking these as roots, they define d maximal subtrees of T, the jth such tree containing r_j further vertices of S. For v, we answer the edge that goes to a child that has the minimal r_j value. This defines an answer to every vertex in round i.

We claim that this way at most \log'_d vertices from round i will be on the final path P. Suppose that there are x vertices from round i that are on P. In reverse order we denote them by $q_0, q_1, \ldots, q_{x-1}$, where q_{x-1} is the one closest to the root. Define Q_j as the maximal subtree of T having q_j as its root. We prove by induction on j that $|Q_j \cap S| \ge 1 + d + d^2 + \ldots + d^j$. For j = 0 this is obvious. For a general j we define the trees $R_1, \ldots R_d$, where R_l is the maximal tree having q_j 's l'th child as its root. Wlog. assume that the edge of P at q_j goes to the root of R_1 . Now by the way the adversary answers, we know that $r_l = R_l \cap S$ is minimal for l = 1 and as R_1 contains the vertex q_{j-1} and everything below it, by induction $|R_1 \cap S| \ge |Q_{j-1} \cap S| \ge 1 + d + d^2 + \ldots + d^{j-1}$. Summing this up, $|Q_i \cap S| = 1 + |R_1 \cap S| + |R_2 \cap S| + \ldots + |R_d \cap S| \ge 1 + d(1 + d + d^2 + \ldots + d^{j-1}) = 1 + d + d^2 + \ldots + d^j$ as claimed. Finishing the proof, $k = |S| \ge |Q_{x-1} \cap S| \ge 1 + d + d^2 + \ldots + d^{x-1}$ implies that $x \le \log'_d k$.

As $1+2+2^2+\ldots+2^{i-1}=2^i-1$, for d=2 the formula in Theorem 3 can be simplified:

Corollary 3.1. For a complete binary tree on n levels, $pa_k(T_2(n)) = si_k(T_2(n)) = \lceil n/\lfloor \log_2(k+1) \rfloor \rceil$

We remark that for general trees, the obvious algoritm is not always the best possible. For example take the tree that starts from the root with a path $p_1p_2...p_n$ of length n, each p_i having one further child except p_n , where there is a complete n level binary tree with root p_n . Now for k=2 the obvious algorithm asking always close to the root, increases P_{max} by 2 in

each round until reaching p_n , then while processing the binary tree, in each round it can only increase P_{max} by 1, thus it finishes approximately in 3n/2 rounds. On the other hand, if in each round we ask one vertex from the path and one vertex from the remainder of the binary tree, in each round the path shortens by one and the binary tree has one less levels. Thus in n rounds we can finish with both parts.

4 Search in pyramid paths

The pyramid graph Py(n) is a directed graph defined in the following way. Py(n) has N = n(n+1)/2 vertices on n+1 levels, for $1 \le i \le n+1$ the *i*th level having *i* vertices $v_{i,1}, v_{i,2} \dots v_{i,i}$, and from every vertex $v_{i,j}$ where $1 \le i \le n$ and $1 \le j \le i$, there is a left outgoing edge going to $v_{i+1,j}$ (its left child) and a right outgoing edge going to $v_{i+1,j+1}$ (its right child). Py(n) has one root on the top, $v_{1,1}$ and n+1 sinks on the bottom, the vertices on the (n+1)th level.

Let us suppose we are given a one dimensional random walk and we want to find either the endpoint, or the whole walk. It is not obvious what search model makes sense here. If we can ask which way it goes in each step, then clearly we need to ask every step, and that is enough, the order of the questions do not matter. But suppose that a question is the following: which direction does the walk go at the *i*th step if it is in the *j*th position?

Clearly it is equivalent to our model on the pyramid graph.

Observation 4.1. For arbitrary k and n we have

- (a) $pa_1(Py(n)) \le pa_k(Py(n)),$
- (b) $pa_{k+1}(Py(n)) \le pa_k(Py(n)),$
- (c) $si_{k+1}(Py(n)) \le si_k(Py(n))$.

It is trivial that if k < N then in both cases we need at least 2 rounds. I.e. the non-adaptive version (having 1 round) of both problems needs N queries. Indeed, for k < N there is a vertex that we did not ask and so the adversary can answer in a way that the path leads to this vertex, everything below this vertex is in a left state and so the state of this non-asked vertex would determine the path and the sink as well.

The fully adaptive version of the problem is again pretty simple.

Claim 4.2.
$$si_1(Py(n)) = pa_1(Py(n)) = n$$
.

It follows from Theorem 1 but we also give a simple proof specific to pyramid paths.

Proof. In n rounds it is easy to determine the path and its sink. First we ask the root and then according to the answer, its left or right outgoing neighbor. We continue this way, in the ith round determining the ith edge of the path, finally asking a vertex on the nth level, thus determining the whole path.

Suppose now that we asked less than n questions, then there is a level $i \leq n$, from which we did not ask any vertex. The adversary answers always left and also at the end he tells us that any vertex not on level i is in left state. Thus we know which vertex of level i is on the path, and the state of this vertex (which we don't know) would determine the path and also its sink.

Conjecture 4.3.
$$si_{s_l}(Py(n)) = pa_{s_l}(Py(n)) = \lceil n/l \rceil$$
 if $s_l = 1 + 2 + ... + l$ for some l .

The upper bound holds by a simple algorithm:

Claim 4.4.
$$si_{s_l}(Py(n)) \le pa_{s_l}(Py(n)) \le \lceil n/l \rceil$$
 if $s_l = 1 + 2 + ... + l$ for some l .

Proof. The algorithm is recursive. In the first round we ask the s_l vertices that are on the first l level, thus we will know the first l edges of the path and also we know the vertex u on the (l+1)st level that is on the path. Now take the new pyramid graph with n levels with root u, by recursion we can find the path here in $\lceil (n-1)/l \rceil = \lceil n/l \rceil - 1$ rounds. This path together with the first l edges gives the path we were looking for in the original pyramidal graph and we had $1 + \lceil n/l \rceil - 1 = \lceil n/l \rceil$ rounds as needed. To start the recursion we need that if $n \leq l$ then one round is enough. This is trivially true as in one round we can ask all the vertices that are not sinks, and so we can determine the path.

The main result of this section is the following. We give a general lower bound that verifies Conjecture 4.3 for l = 2 (i.e. $k = s_l = 3$) and solves the case k = 2.

Theorem 4. For arbitrary
$$k$$
, $pa_k(Py(n)) \ge si_k(Py(n)) \ge \lceil \frac{2}{k+1}n \rceil$.

Proof. We give two different proofs. While processing an algorithm which finds the path/sink, there is always a maximal partial pyramid path P_{max} that we know from the answers until now, i.e. the path determined by the state of the already known vertices (note that P_{max} is changing by time). The basic idea in both proofs is that in each round there is only one question which immediately makes P_{max} longer by one and for the rest of the questions, only pairs of them can determine one more edge in P_{max} . In both proofs the adversary has the following answering scheme. In each round he answers for the k asked vertices in reverse order of their height i.e. he first answers for the one which is on a level with a biggest index (if there are more asked vertices on the same level, then their order does no matter), etc.. This way at most one vertex per round is the endvertex of P_{max} when it is asked.

First proof of the lower bound. If a vertex is not an endvertex of P_{max} then we just answer left. If the asked vertex v is an endvertex of P_{max} then we do the following. Let v's left child be u and its right child be w. Compute the length l_u of the path starting from u determined by the already known states of vertices (when we reach a vertex with unknown state, that's the end of the path, eg. it may be already u if we don't know u's state). Similarly, the length of the path starting at w is l_w . Now we answer left for the state of v if $l_w \leq l_u$ and right otherwise i.e. we choose to go in the direction where the continuation of the path will be shorter.

When analyzing this method we just concentrate on the vertices which at the stage when they are asked, are the endpoints of the current P_{max} . As already mentioned, we consider only one such point in each round. After our answer to a vertex v the path gets longer by at least one. If by more than one, then wlog. we have chosen u, its left child and so $l_u \leq l_w$, where w is its right child. As for non-endpoints of P_{max} we always answer left and the two paths starting at u and w contain only such vertices, they are completely disjoint. I.e. for all but one edge of the new P_{max} determined in this round, we found two vertices for which the state was asked already. As P_{max} is increasing, in each round we find new such pairs of vertices. It is easy to see that doing this the sink is determined if and only if the pyramid path is determined as well. Suppose now that after m rounds the whole pyramid path is determined, i.e. all n edges of it. In each round there was at most one endvertex of P_{max} asked, which means that at most m edges were determined by them and for the rest of the edges we found two asked vertices for each. Thus together there where at least m + 2(n - m) = 2n - m questions. Thus we had at least $m \geq (2n - m)/k = (2n - m)/k$ rounds which implies $m \geq \frac{2}{k+1}n$ as needed.

Second proof of the lower bound. If a vertex v is not an endvertex when it is asked then we check if there is another vertex v' on the same level with known state. If there is, then we give the same answer for v as we gave for v' and additionally we tell that the state of every vertex on this level is the same. Otherwise we give an arbitrary answer and also say that either there will be one more asked vertex in this level or we will avoid v. If v is an endvertex of P_{max} , then we determine the first level under it for which we did not tell the state of every vertex on that level. On this level either there is no asked vertex, then our answer to v is arbitrary or there is a vertex q which was already asked. Now either answering left or right to v will make sure that the endvertex of the new P_{max} is on the same level as q but a different vertex. Thus the question when we asked q became useless, wlog. we can assume that there is no such q. This way questions which were endvertices of P_{max} determine one edge in P_{max} and pairs of the rest of the vertices determine a whole level, i.e. one edge in P_{max} . It is easy again to see that doing this the sink is determined if and only if the pyramid path is determined as well. The same computation as in the first proof yields the desired lower bounds.

Corollary 4.5.
$$pa_2(Py(n)) = si_2(Py(n)) = \lceil \frac{2}{3}n \rceil$$
. $pa_3(Py(n)) = si_3(Py(n)) = \lceil \frac{1}{2}n \rceil$.

Proof. Theorem 4 implies $si_2(Py(n)) \ge \lceil \frac{2}{3}n \rceil$ and $si_3(Py(n)) \ge \lceil \frac{1}{2}n \rceil$. We now need to give algorithms for finding the paths, that achieve these bounds. For k=3 the algorithm in Claim 4.4 can be applied. For k=2 first we ask the root $v_{1,1}$ and $v_{3,2}$. Wlog. the root is in left state. Now in the second round we ask $v_{2,1}$ and $v_{3,1}$. After these two rounds we will know the first 3 edges of the path and then we can proceed by recursion (taking the endvertex of this 3 long path as the new root).

Pyramid graphs can be easily generalized to d-dimensions, see eg. the paper of Sun et al. [4] In this paper a pyramid graph is represented on the non-negative part of the 2-dimensional grid with the origo being its root. In a similar way a d-dimensional pyramid graph is a part of the d-dimensional grid. In the following we give a generalization of pyramid paths that includes the d-dimensional pyramid path. A generalized pyramid graph $GPy_d(n)$ is a directed graph having the following properties. $GPy_d(n)$ has its vertices on n+1 levels such that the first level has one source vertex and the last level contains only sinks. From any vertex v_i on level $i \le n$, there are d outgoing edges to level i+1, and between each two levels i and i+1 there is a matching L_i that matches level i to level i+1. As a consequence, on each non-first level there are at least d vertices.

Theorem 5. For any generalized pyramid graph $GPy_d(n)$ and arbitrary k, $pa_k(Py(n)) \ge si_k(Py(n)) \ge \lceil \frac{d}{k-1+d}n \rceil$.

Proof. Both proofs of Theorem 4 easily generalize to this setting. Here we present a proof using the second method. We refer to the edges of all the L_i 's as left edges. If a vertex v is not an endvertex when it is asked then we check if there are at least d-1 another vertices on the same level with known state. If no, then we give answer left (i.e. the edge from the appropriate matching) and also say that either there will be one more asked vertex in this level or the final path won't go through v. If there are at least d-1 vertices already on this level with known state, then we again give answer left for v and additionally we tell that the state of every vertex on this level is left. If v is an endvertex of P_{max} which is on level i, then we determine the first level j under it for which we did not tell the state of every vertex on that level. If on this level there are at most d-1 asked vertices, then our answer to v is such that the new endvertex of P_{max} is a vertex on level j that was not yet asked. This can be done, as the at least d different choices for the state of v all yield to different endvertices on all the levels from i to j, as on every level every known state is a matching edge and the matching edges never go to the same vertex.

It is again easy to conclude that in each round there is only one question which immediately makes P_{max} longer by one and for the rest of the questions, only d-tuples of them can determine one more edge in P_{max} . Simple computation gives the lower bound $si_k(GPy_d(n)) \ge \frac{d}{k-1+d}n$. \square

We also remark that Theorem 5 cannot be improved as there are generalized pyramid graphs for every d where equality holds. Indeed, take the n+1 level generalized pyramid graph having d vertices on each non-first level for which every non-sink vertex is connected to every vertex on the next level. This graph is uniquely determined by n and d. Now in this graph an obvious algorithm is to ask in each round the endvertex of P_{max} and as many complete levels under it as possible. This way in each round P_{max} gets longer by $1 + \lfloor (k-1)/d \rfloor = \lfloor (k-1+d)/d \rfloor$, thus we can determine the path in at most $\lceil n/(\lfloor \frac{k-1+d}{d} \rfloor) \rceil$ rounds, If k-1 is divisible by d then this upper bound matches the lower bound of Theorem 5.

5 Remarks

Let us consider again the problem of $pa_k(G)$ and $si_k(G)$ for directed acyclic graphs. Obviously the same preprocessing as in the case k = 1, replacing G by G' or G'', is useful in general. From now on we suppose that G does not have vertices with outdegree 1 (which is always true for G' and G'').

One could ask how far si(G) and $si_k(G)$ (or pa(G) and $pa_k(G)$) can be. Obviously $si(G) \le ksi_k(G)$ as the same at most $ksi_k(G)$ questions which were used to find the sink in the case of $si_k(G)$ could be used one-by-one to find the sink in the case of si(G). An example where this bound is achieved is the graph H_l consisting of a directed path of length kl with each vertex on the path having another out-neighbor, which is a sink. More precisely let x_1, \ldots, x_{kl+1} be vertices of H_l such that $x_ix_{i+1} \in E$ for every $i \le kl$. Additionally, every x_i with $i \le kl$ has a child, which is a sink. One can easily see that in the worst case (that can be forced by the adversary) the sink is found if and only if x_1, \ldots, x_{kl} have been asked, hence $si(H_l) = kl$ and $si_k(H_l) = l$. The same is true if we want to find the path.

An example where si(G) and $si_k(G)$ (or pa(G) and $pa_k(G)$) are close to each other is the complete k-ary tree from Section 3. It follows easily from Theorem 3 that $si(T_k(n)) = si_k(T_k(n)) = pa(T_k(n)) = pa_k(T_k(n)) = n$.

More generally, one could ask how far $si_m(G)$ and $si_k(G)$ (or $pa_m(G)$ and $pa_k(G)$) can be for any constants $m \leq k$. Obviously $si_m(G) \geq si_k(G)$, and similarly to the arguments used in the case m = 1, $si_m(G) \leq \lceil k/m \rceil si_k(G)$.

One can easily construct a graph where this bound is achieved. We just mention the main ideas without details. Let us consider a k-ary tree with l levels and replace each vertex with a copy of H_1 . The trivial algorithm is to go through the copy of H corresponding to the source of the tree, then in the appropriate child of it, and so on. This gives $si_k(G) \leq pa_k(H') \leq l$ and $si_m(H') \leq \lceil k/m \rceil l$. On the other hand the method of the adversary can be the following: if a vertex is asked and it is not in the upper-most copy of H_1 , the path won't even go into that copy of H_1 which contains this vertex. It shows that there is equality in the previous inequalities.

Summarizing:

Claim 5.1. For arbitrary $m \leq k$ $si_m(G) \leq \lceil k/m \rceil si_k(G)$ and there are infinitely many G graphs for which equality holds.

For larger k, one can easily improve the trivial (and optimal) algorithm we mentioned in Section 2 for k = 1. At first we ask the source, all its out-neighbors, every vertex which can be reached from the source in a path of length two and so on. If there is an i for which we cannot ask every vertex which can be reached from the source in a path of length i, then we ask as many as we can, chosen arbitrarily. Then the answers show the beginning of the path $P(E^*)$. We repeat this procedure starting with the last vertex which is surely in $P(E^*)$. This algorithm (let us call it Algorithm A) finds the sink and the path too, using at most l(G) questions.

Clearly a smarter algorithm cannot be more than k times faster than this trivial one for a graph (without vertices of outdegree 1), as it would mean $si_k(G) < l(G)/k = si(G)/k$.

Now for any k we show a graph G and algorithm which can achieve this bound (depending on how the arbitrary vertices are chosen in the trivial algorithm). Let x_1, \ldots, x_{kl+1} be vertices of the graph such that $x_i x_{i+1} \in E$ for every $i \leq kl$. Additionally, every x_i with i < kl has k-1 children, each of them having two children, and x_{kl} has k-1 additional children. These additional vertices are all distinct, hence the graph is a tree. Algorithm A asks x_1 and k-1 of its children in the first turn. It is possible that it does not ask x_2 . Suppose the path $P(E^*)$ goes to x_{kl+1} . If the arbitrary vertices are always chosen the worst possible way, than kl turns are needed (even if they are chosen smarter, at least kl/2 turns are needed).

However, consider the following Algorithm B. At the first turn we ask x_1, \ldots, x_k . If the path $P(E^*)$ does not go to x_{k+1} , then we need to ask one more questions to finish the algorithm, otherwise we continue with x_{1+1}, \ldots, x_{2k} , and so on. One can easily see that Algorithm B finishes after at most l turns.

Summarizing (we denote by $si_A(G)$ the number of steps in which Algorithm A finds the sink):

Claim 5.2. $si_k(G) \geq si_A(G)/k$ and there are infinitely many G graphs for which equality holds.

Also, as we noted before, our proof that pa(G) = l(G), if there is no vertex with outdegree 1, can be interpreted even for graphs containing cycles, yet we could not prove such a claim for si(G) if G contains a cycle. Even the question is not clear in this case. We could ask for the sink or cycle where the flow ends, analogously to the acyclic case. On the other hand if we want to know every edge of the cycle, it is more similar to pa(G). A possible goal could be to determine the sink or the last vertex before/after creating a cycle in the flow.

Problem 5.3. Give an efficient algorithm to determine in this sense si(G) if G contains a cycle.

The most interesting open problem is still to determine $pa_k(Py(n))$ for every k and n or at least to prove Conjecture 4.3. Further, our results suggest that the following might be true.

Problem 5.4. Is it true that $pa_k(Py(n)) = si_k(Py(n))$ for every k and n?

In a paper of Sun et al. [4] a very similar problem was investigated. In their version of the problem, in one question we can ask for a vertex if it is on the path or not. Let us call the minimal number of questions for this version $pa'_k(G)$ and $si'_k(G)$. For pyramid graphs, the completely adaptive (i.e. one questions per round) version, similarly to our problem, we need n rounds. However, they do not regard the version when we can ask more questions per round. It is trivial that a question of our kind can be emulated by 3 questions of their kind (asking the vertex and also its 2 outgoing neighbors), thus $pa_k(Py(n)) \leq 3pa'_k(Py(n))$ and $pa(Py(n)) \leq 3si'_k(Py(n))$. It would be interesting to know more about these two new functions.

What Sun et al. investigate is that for pa' and si', eg. for pyramid graphs algorithms using randomization can find the path much faster than deterministic ones. This might show one major difference between these two sets of problems, as in the version we regard, randomization does not seem to help much. One possible intuition behind this difference is that a left/right answer just gives a relative information, which might be completely useless to determine our path, whereas in their case any answer gives some information about the path, i.e. whether it goes through that vertex or not.

Also, in their paper this was a major tool to give bounds to various Local Search Problems. It would be interesting to see whether our version has similar theoretical applications.

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